

Stage-wise Training Strategy for **Multiple Mental Disorder Detection**

FOCUSED KAN IN TRANSFORMERS

Team 5

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Introduction

- 23.1% of U.S. adults experienced AMI in 2022
- 3% having more than one mental illness concurrently.
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Mental disorders significantly affect daily life but are **hard to detect** due to subjective and nuanced symptoms. Social media data, with **spontaneous expressions** of emotions and symptoms, mitigate biases from face to face interactions.

Challenges of training a multi-task model:

- Limited Multi-labeled Datasets
- Unique linguistic & contextual features of each disorder
- Biases in Different Task Difficulties

We propose: **Stage-wise Training Strategy using Transformer with Kan**

Dataset_(StageTraining) *From Reddit Pushshift Dump (2005-2023)*

Case Subreddit

- **r/depression**
- **r/Anxiety**
- **r/OCD**
- **r/bipolar**
- **r/ptsd**

Control Subreddit

- **r/ChangeMyView**
- **r/NoStupidQuestions**
- **r/Showerthoughts**
- **r/books**
- **r/movies**
- **r/CasualConversation**

Data Preprocessing

posts marked as null or empty (e.g., posts labeled as "[removed]" or "[deleted]" or those containing only URLs) were removed, etc.

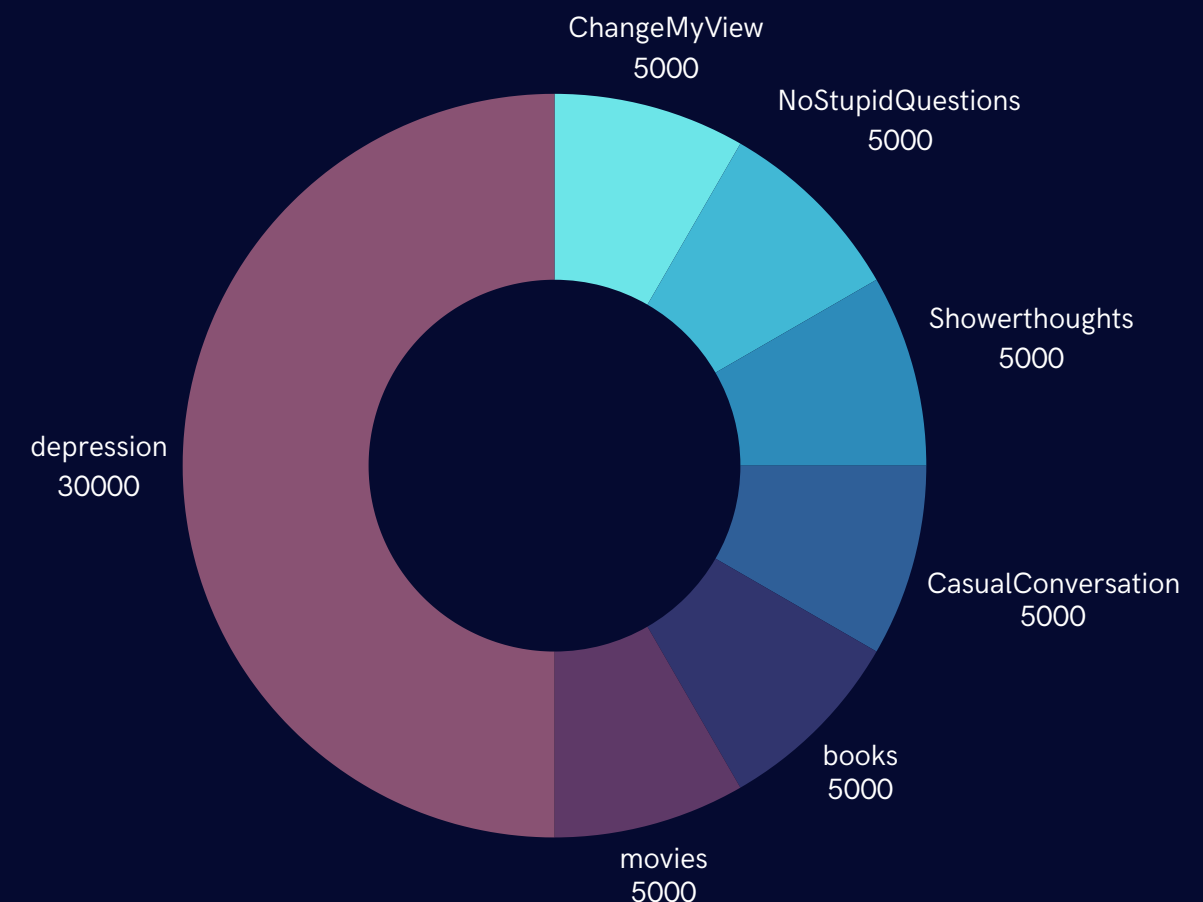


After preprocessing, a subset of:

- Case: **30,000** usable posts
- Control: **25,000** usable posts

Was selected for analysis

Split 80-20 for training and testing



Dataset (Finetuning)

From Kaggle and Chat-GPT

a Kaggle dataset:
Includes texts tagged
with statuses:

“Normal”,
“Depression”,
“Suicidal”,
“Anxiety”,
“Stress”,
“Bi-Polar”,
“Personality Disorder”.

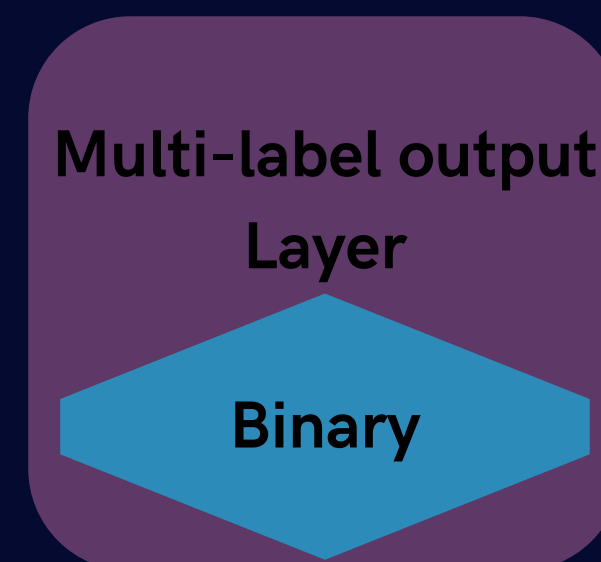
Data Annotation

Map to our 5 target conditions
using GPT 4o

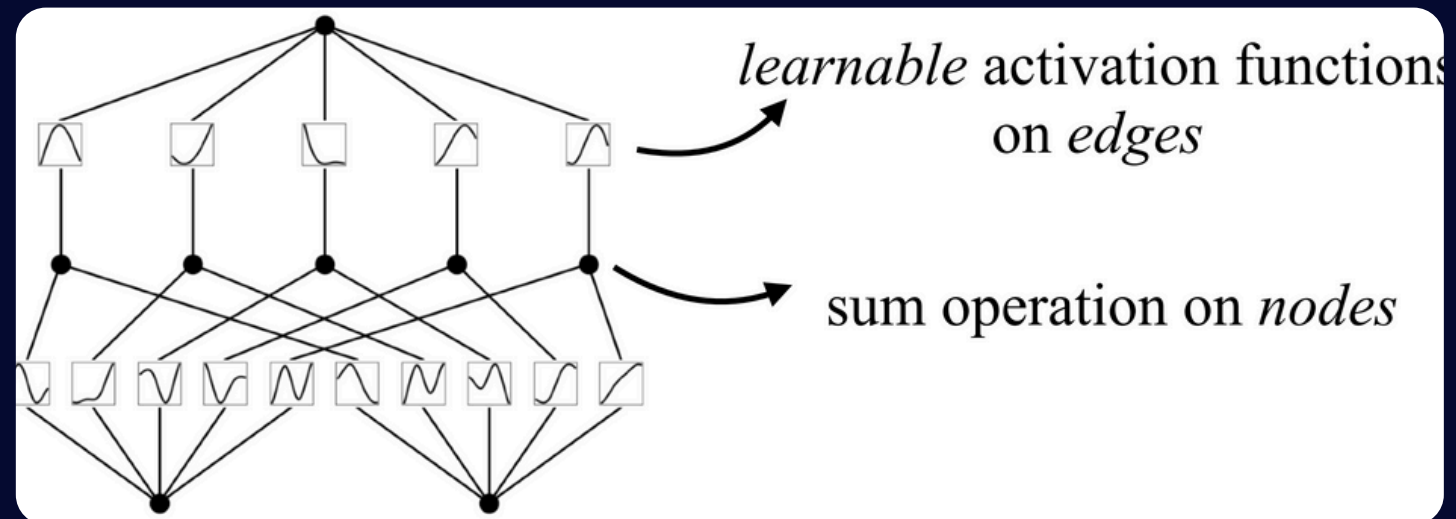


This dataset contains:
53,044 posts, with **7.42%**
**annotated with two or
more condition labels**

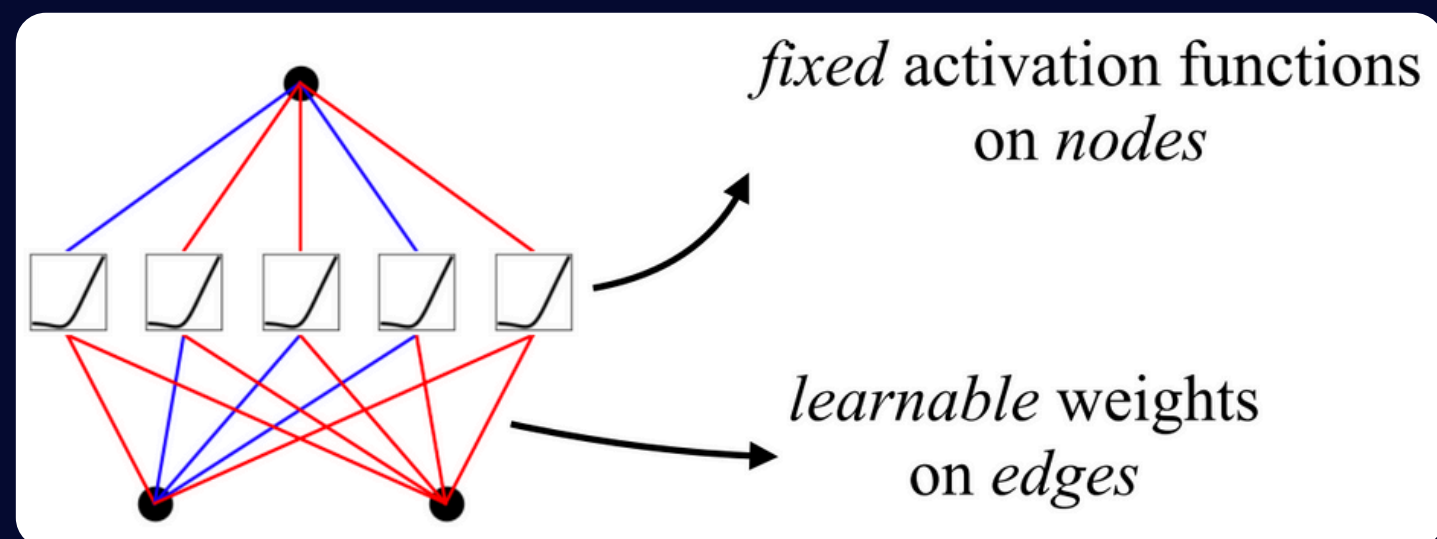
Text	Depression	Anxiety	bipolar	OCD	PTSD
...	1	0	0	0	1
...	0	0	0	1	0
...	1	1	1	0	0
(Normal)	0	0	0	0	0



Method



Kolmogorov-Arnold Network



Multi-Layer Perceptron

KAN:

more flexible

spline-parametrized univariate functions;
dynamically adapt activation patterns.

Multi-Stage Training Strategy



Method (2)

Multi-stage training strategy

1. mental disorder datasets - binary classification tasks

enter a new stage when the model is optimized on current data

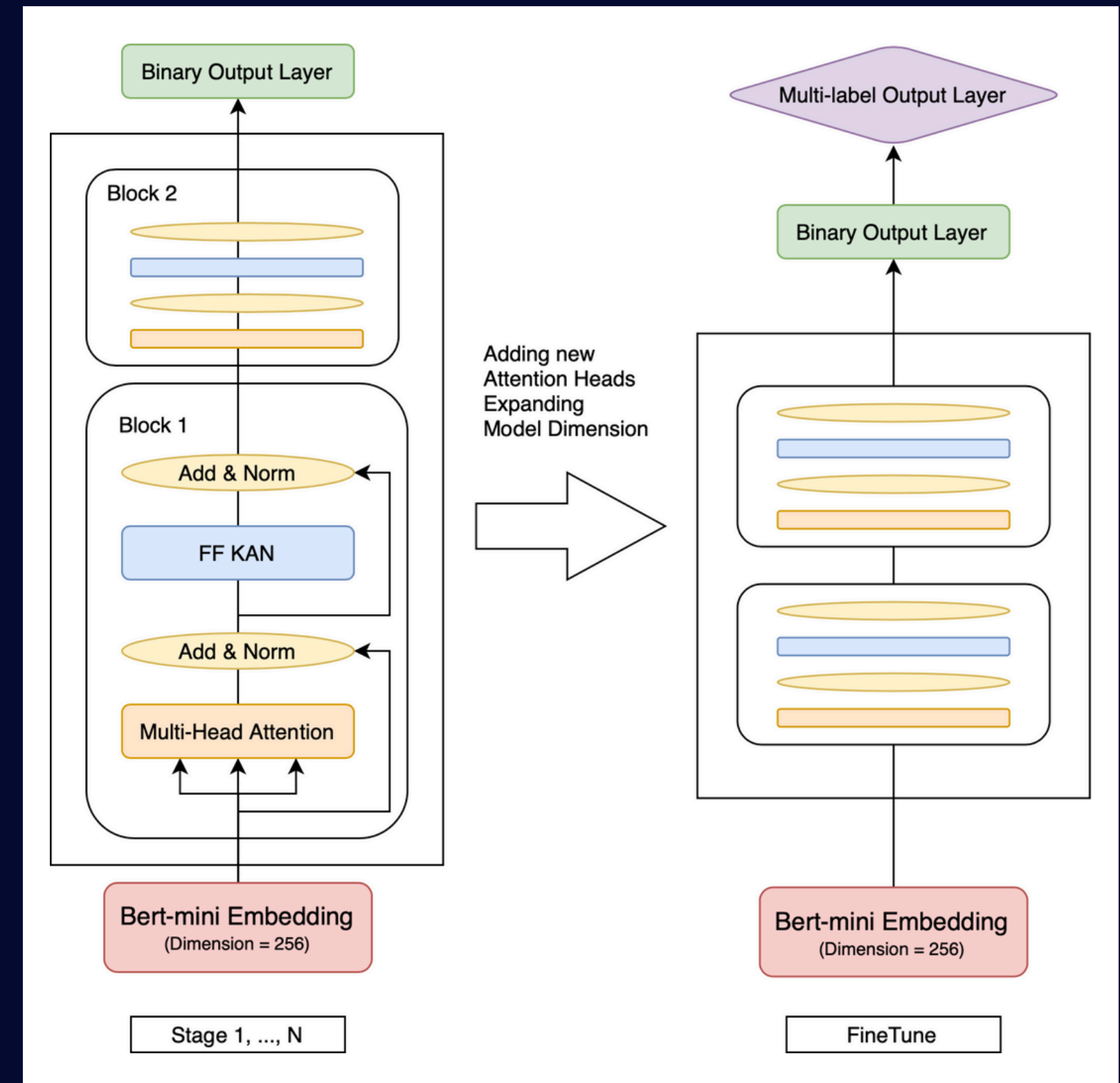
2. **FREEZE**: freeze an attention head once weight changes fall below a threshold during the training process

3. Add new heads: ensure the model has enough capacity to learn new information

Fine-tuning on multi-label mixed Dataset

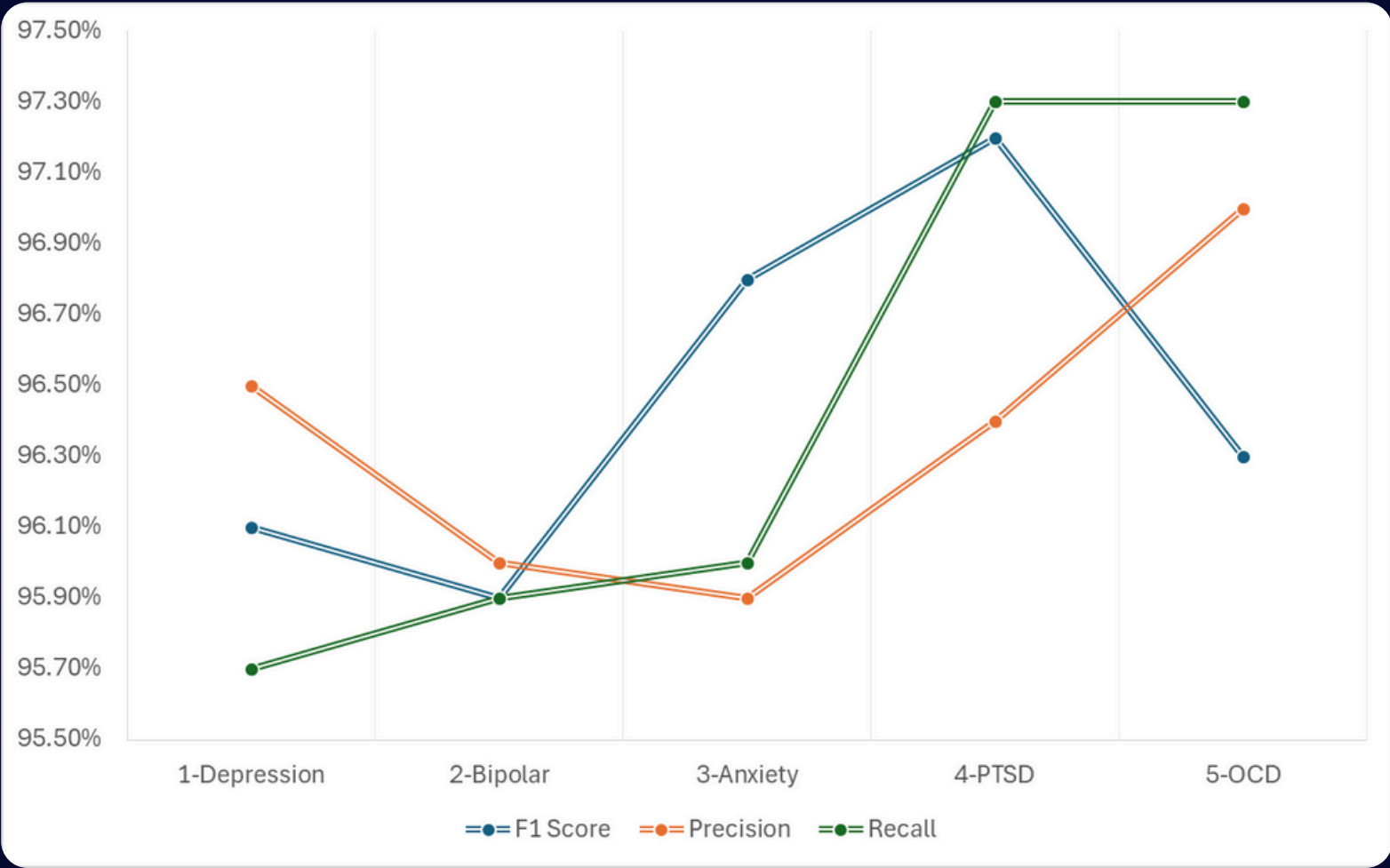
to adjust the model for the new multi-label task

to modify the new output layer



Stage-wise Results

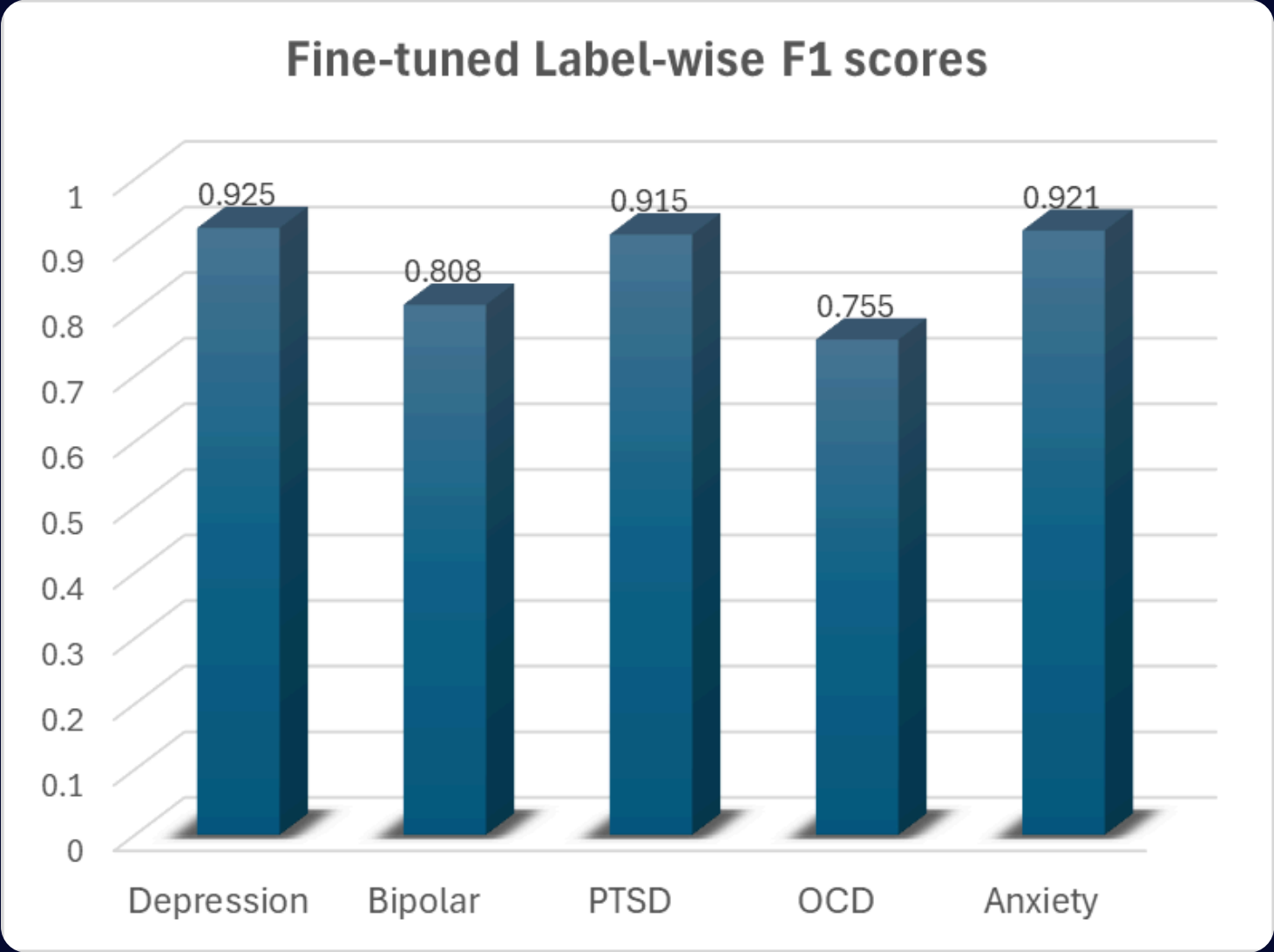
Stage	Train Loss	Validation loss	Recall	Precision	F1
1-Depression	0.208	0.241	0.957	0.965	0.961
2-Bipolar	0.052	0.15	0.959	0.96	0.959
3-Anxiety	0.052	0.159	0.959	0.96	0.968
4-PTSD	0.052	0.112	0.973	0.964	0.972
5-OCD	0.05	0.101	0.973	0.97	0.963



Fine-tuned Results

Label-wise F1 scores	
Depression	0.925
Bipolar	0.867
PTSD	0.915
OCD	0.834
Anxiety	0.933

Overall Model Performance		
Macro F1	Micro F1	EM
0.865	0.915	0.891



“Some days I feel like I'm invincible, bursting with energy and confidence, but other days, I'm drowning in my thoughts, haunted by memories I can't escape, and overwhelmed by sadness so deep it feels like I'll never climb out.” (depression + anxiety)

Depression	Bipolar	PTSD	OCD	Anxiety
0.6262	0.0007	0.0009	0.0007	0.3714

Takeaways

Model Strategies

- KAN-based Transformers
- Stage-Wise Training
- Dynamic Attention Head Freezing
- Transfer Knowledge
- Guided Training
- Continuous Learning

Model Strengths:

- Interpretability
- Task Interference
- Modular Learning
- Catastrophic Forgetting
- Shared Feature Integration
- Parameter Efficiency
- Strong Performance
- Scalability

**Solving
Comorbidity!**